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BOX: PATENT APPLICATION

Assistant Commissioner for Patents
Washington, D.C. 20231

Re: Application of Hyun-doo SHIN, Yang-lim CHOI and B.S. MANJUNATH
METHOD FOR ANALYZING TEXTURE OF DIGITAL IMAGE
Our Reference: Q56074

Dear Sir:

Attached hereto is the application identified above including the specification, claims and five (5) sheets of drawings. The requisite U.S. Government Filing Fee, executed Declaration and Power of Attorney and Assignment will be submitted at a later date.

The Government filing fee is calculated as follows:

Total Claims	42 - 20 =	22 x \$18 =	\$ 396.00
Independent Claims	19 - 3 =	16 x \$80 =	\$ 1,280.00
Base Filing Fee	(\$710.00)		\$ 710.00
Multiple Dep. Claim Fee	(\$260.00)		\$ 000.00
TOTAL FILING FEE			\$ 2,386.00

Priority/Benefit is claimed from:

U.S. Provisional Application

Filing Date

60/157,077

October 1, 1999

Since the anniversary of the priority date fell on a Sunday, the filing of this application on Monday, October 2, 2000, is sufficient to obtain the benefit of priority.

Respectfully submitted,
SUGHRUE, MION, ZINN, MACPEAK & SEAS
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METHOD FOR ANALYZING TEXTURE OF DIGITAL IMAGE

BACKGROUND OF THE INVENTION

1. Field of the Invention

The present invention relates to a method for analyzing of the texture of a digital image, and more particularly to a method for analyzing the texture of a digital image using a texture descriptor. This application claims the priority benefit of provisional application Serial No. 60/157,077, entitled "METHOD FOR ANALYZING OF STILL IMAGE TEXTURE", filed on October 1, 1999.

2. Description of the Related Art

FIG. 1 illustrates a conventional digital image texture analyzing method.

Referring to FIG. 1, filtered images are obtained by performing a predetermined filtering method (step 102), and the mean values and variances of the pixel values of the filtered images are used as texture descriptors of the images (steps 104 and 106). The texture descriptors indicate an analysis result of image texture.

However, according to the conventional digital image texture analyzing method, even when the textures of an image are similar to those of another image, in the case where one image is rotated, enlarged or reduced relative to the other image, the textures of the two images are analyzed to be different.

SUMMARY OF THE INVENTION

It is a first object of the present invention to provide a digital image texture analyzing method for analyzing and comparing textures of images more accurately.

It is a second object of the present invention to provide a computer-readable recording medium for storing program codes for performing the digital image texture analyzing method for accomplishing the first object of the present invention.

It is a third object of the present invention to provide a digital image texture analyzing apparatus for performing the digital image texture analyzing method for accomplishing the first object of the present invention.

It is a fourth object of the present invention to provide a digital image searching method for searching for an image using a texture descriptor indexed by the digital image texture analyzing method for accomplishing the first object of the present invention.

5 It is a fifth object of the present invention to provide a digital image texture analyzing method for accurately analyzing the similarity of textures of two images.

It is a sixth object of the present invention to provide a computer-readable recording medium for storing program codes for performing the digital image texture analyzing method for accomplishing the fifth object of the present invention.

10 It is a seventh object of the present invention to provide a digital image texture analyzing apparatus for performing the digital image texture analyzing method for accomplishing the fifth object of the present invention.

15 It is an eighth object of the present invention to provide a digital image texture analyzing method for accurately analyzing that textures of two images are similar even when one image is rotated relative to the other image.

It is a ninth object of the present invention to provide a computer-readable recording medium for storing program codes for performing the digital image texture analyzing method for accomplishing the eighth object of the present invention.

20 It is a tenth object of the present invention to provide a digital image texture analyzing apparatus for performing the digital image texture analyzing method for accomplishing the eighth object of the present invention.

It is an eleventh object of the present invention to provide a digital image texture analyzing method for accurately analyzing that textures of two images are similar even when one image is enlarged or reduced relative to the other image.

25 It is a twelfth object of the present invention to provide a computer-readable recording medium for storing program codes for performing the digital image texture analyzing method for accomplishing the eleventh object of the present invention.

30 It is a thirteenth object of the present invention to provide a digital image texture analyzing apparatus for performing the digital image texture analyzing method for accomplishing the eleventh object of the present invention.

It is a fourteenth object of the present invention to provide a digital image texture analyzing method for accurately analyzing that textures of two images are similar even when one image is rotated, enlarged or reduced relative to the other image.

Accordingly, to achieve the first object, there is provided a digital image texture analyzing method including the step of obtaining a texture descriptor including the mean of the pixel values of an original image as a texture feature.

Here, the texture feature may further include the variance of the pixel values of the original image.

Also, the digital image texture analyzing method may include the step of obtaining a texture descriptor including the variance of the pixel values of an original image as a texture feature.

According to another aspect of the invention, there is provided a digital image texture analyzing method including the steps of (a) obtaining the mean (μ_0) and variance (σ_0) of the pixel values of an original image, (b) obtaining $m \times n$ filtered images by filtering the original image using predetermined filters, each having a unique combination of one of m scales and one of n orientations, where m and n are predetermined positive integers, (c) calculating the means ($\mu_{11}, \mu_{21}, \dots, \mu_{mn}$) and variances ($\sigma_{11}, \sigma_{21}, \dots, \sigma_{mn}$) of the respective filtered images, and (d) obtaining a texture descriptor having the mean (μ_0) and variance (σ_0) of the pixel values of the original image obtained in the step (a), and the means ($\mu_{11}, \mu_{21}, \dots, \mu_{mn}$) and variances ($\sigma_{11}, \sigma_{21}, \dots, \sigma_{mn}$) of the respective filtered images obtained in the step (c), as texture features.

To achieve the second object of the present invention, there is provided a computer-readable recording medium for storing program codes for performing a digital image texture analyzing method including the steps of (a) obtaining the mean (μ_0) and variance (σ_0) of the pixel values of an original image, (b) obtaining $m \times n$ filtered images by filtering the original image using predetermined filters, each having a unique combination of one of m scales and one of n orientations, where m and n are predetermined positive integers, (c) calculating the means ($\mu_{11}, \mu_{21}, \dots, \mu_{mn}$) and variances ($\sigma_{11}, \sigma_{21}, \dots, \sigma_{mn}$) of the respective filtered images, and (d) obtaining a texture descriptor having the mean (μ_0) and variance (σ_0) of the pixel values of the original image obtained in the step (a), and the means ($\mu_{11}, \mu_{21}, \dots, \mu_{mn}$) and variances ($\sigma_{11}, \sigma_{21}, \dots, \sigma_{mn}$) of the respective filtered images, obtained in the step (c), as texture features.

To achieve the third object of the present invention, there is provided a digital image texture analyzing apparatus including a mean/variance calculating unit for calculating the mean and variance of the pixel values of an original image, and a texture descriptor setting unit for setting the mean and variance as texture descriptors.

Also, the apparatus may further include a filtering unit for filtering the original image using predetermined filters, each having a unique combination of one of m scales and one of n orientations, where m and n are predetermined positive integers, to obtain $m \times n$ filtered images, wherein the mean/variance calculating unit obtains the mean (μ_0) and variance (σ_0) of the pixel values of the original image and the means ($\mu_{11}, \mu_{21}, \dots, \mu_{mn}$) and variances ($\sigma_{11}, \sigma_{21}, \dots, \sigma_{mn}$) of the respective filtered images, and wherein the texture descriptor setting unit obtains a texture descriptor having the mean (μ_0) and variance (σ_0) of the pixel values of the original image, and the means ($\mu_{11}, \mu_{21}, \dots, \mu_{mn}$) and variances ($\sigma_{11}, \sigma_{21}, \dots, \sigma_{mn}$) of the respective filtered images, as texture features.

To achieve the fourth object of the present invention, there is provided a digital image searching method including the step of searching for an image having a similar texture descriptor to a query image using a texture descriptor having the mean and variance of the pixel values of an original image as texture features.

To achieve the fifth object of the present invention, there is provided a digital image texture analyzing method for evaluating the similarity of textures of two arbitrary digital images by obtaining a matching metric between the two images, the method including the step of calculating the matching metric including absolute differences between the means of the pixel values of the two arbitrary digital images.

Also, the matching metric may further include absolute differences between the variances of the pixel values of the original image.

According to another aspect of the present invention, there is provided a digital image texture analyzing method for evaluating the similarity of textures of two arbitrary digital images by obtaining a matching metric between the two images, the method including the step of calculating the matching metric including absolute differences between the variances of the pixel values of the two arbitrary digital images.

Alternatively, the digital image texture analyzing method may include the steps of (a) obtaining $m \times n$ filtered images with respect to two arbitrary images using predetermined filters, each having a unique combination of one of m scales and one of n orientations, where m and n are predetermined positive integers, and (b)

5 calculating the matching metric defined by:

$$d(i, j) = \sum_{m,n} d_{m,n}(i, j) + b$$

with respect to the original image and the $m \times n$ filtered images, where the mean and variance of the pixel values of the respective images are μ and σ , the mean and variance of the pixel values of the original image are μ_0 and σ_0 ,

$$b = \left| \frac{\mu_0^{(i)} - \mu_0^{(j)}}{\alpha(\mu_0)} \right| + \left| \frac{\sigma_0^{(i)} - \sigma_0^{(j)}}{\alpha(\sigma_0)} \right|, \quad d_{m,n}(i, j) = \left| \frac{\mu_{m,n}^{(i)} - \mu_{m,n}^{(j)}}{\alpha(\mu_{m,n})} \right| + \left| \frac{\sigma_{m,n}^{(i)} - \sigma_{m,n}^{(j)}}{\alpha(\sigma_{m,n})} \right|, \text{ and } \alpha(\cdot) \text{ is a}$$

10 predetermined scaling function.

To achieve the sixth object of the present invention, there is provided a computer-readable recording medium for storing program codes for performing a digital image texture analyzing method for evaluating the similarity of textures of two arbitrary digital images by obtaining a matching metric between the two images, the method including the steps of (a) obtaining $m \times n$ filtered images with respect to two

15 arbitrary images using predetermined filters, each having a unique combination of one of m scales and one of n orientations, where m and n are predetermined positive integers, and (b) calculating the matching metric defined by:

$$d(i, j) = \sum_{m,n} d_{m,n}(i, j) + b$$

with respect to the original image and the $m \times n$ filtered images, where the mean and variance of the pixel values of the respective images are μ and σ , the mean and variance of the pixel values of the original image are μ_0 and σ_0 ,

$$b = \left| \frac{\mu_0^{(i)} - \mu_0^{(j)}}{\alpha(\mu_0)} \right| + \left| \frac{\sigma_0^{(i)} - \sigma_0^{(j)}}{\alpha(\sigma_0)} \right|, \quad d_{m,n}(i, j) = \left| \frac{\mu_{m,n}^{(i)} - \mu_{m,n}^{(j)}}{\alpha(\mu_{m,n})} \right| + \left| \frac{\sigma_{m,n}^{(i)} - \sigma_{m,n}^{(j)}}{\alpha(\sigma_{m,n})} \right|, \text{ and } \alpha(\cdot) \text{ is a}$$

predetermined scaling function.

To achieve the seventh object of the present invention, there is provided a digital image texture analyzing apparatus for evaluating the similarity of textures of two arbitrary digital images by obtaining a matching metric between the two images, the apparatus including a filtering unit for obtaining $m \times n$ filtered images with respect to two arbitrary images using predetermined filters, each having a unique combination of one of m scales and one of n orientations, where m and n are predetermined positive integers, and a matching metric calculating unit for calculating the matching metric defined by:

$$d(i, j) = \sum_{m,n} d_{m,n}(i, j) + b$$

with respect to the original image and the $m \times n$ filtered images, where the mean and variance of the pixel values of the respective images are μ and σ , the mean and variance of the pixel values of the original image are μ_0 and σ_0 ,

$$b = \left| \frac{\mu_0^{(i)} - \mu_0^{(j)}}{\alpha(\mu_0)} \right| + \left| \frac{\sigma_0^{(i)} - \sigma_0^{(j)}}{\alpha(\sigma_0)} \right|, \quad d_{m,n}(i, j) = \left| \frac{\mu_{m,n}^{(i)} - \mu_{m,n}^{(j)}}{\alpha(\mu_{m,n})} \right| + \left| \frac{\sigma_{m,n}^{(i)} - \sigma_{m,n}^{(j)}}{\alpha(\sigma_{m,n})} \right|, \text{ and } \alpha(\cdot) \text{ is a}$$

predetermined scaling function.

To achieve the eighth object of the present invention, there is provided a digital image texture analyzing method for evaluating the similarity of textures of two arbitrary digital images by obtaining a matching metric between the two images, the method including the steps of (a) obtaining filtered images using predetermined filters having different orientation coefficients with respect to two arbitrary images, and (b) obtaining a matching metric by calculating the minimum value of the sum of absolute differences between the means of the pixel values with respect to an arbitrary filtered image and images filtered by filters having orientation coefficients different from those of filters used for filtering the arbitrary image.

According to another aspect of the present invention, there is provided a digital image texture analyzing method for evaluating the similarity of textures of two arbitrary digital images by obtaining a matching metric between the two images, the method including the steps of (a) obtaining filtered images using predetermined filters having different orientation coefficients with respect to two arbitrary images, and (b) obtaining a matching metric by calculating the minimum value of the sum of

absolute differences between the variances of the pixel values with respect to an arbitrary filtered image and images filtered by filters having orientation coefficients different from those of filters used for filtering the arbitrary image.

To achieve the ninth object of the present invention, there is provided a computer-readable recording medium for storing program codes for performing a digital image texture analyzing method for evaluating the similarity of textures of two arbitrary digital images by obtaining a matching metric between the two images, the method including the steps of (a) obtaining $m \times n$ filtered images with respect to two arbitrary images using predetermined filters, each having a unique combination of one of m scales and one of n orientations, where m and n are predetermined positive integers, and (b) calculating the matching metric defined by:

$$d_{m,n}(i, j) = \min_{1 \leq t \leq K} \left[\sum_{m,n} \left(\left| \frac{\mu_{m,n}^{(i)}}{\alpha(\mu_{m,n})} - \frac{\mu_{m,n \oplus t}^{(j)}}{\alpha(\mu_{m,n})} \right| \right) + \left(\left| \frac{\sigma_{m,n}^{(i)}}{\alpha(\sigma_{m,n})} - \frac{\sigma_{m,n \oplus t}^{(j)}}{\alpha(\sigma_{m,n})} \right| \right) \right]$$

with respect to the $m \times n$ filtered images, where the mean and variance of the pixel values of the respective images are μ and σ , K is a predetermined positive integer representing the number of orientations to be considered, \oplus denotes a modulo shift function, and $\alpha(\)$ is a predetermined scaling function.

To achieve the tenth object of the present invention, there is provided a digital image texture analyzing apparatus for evaluating the similarity of textures of two arbitrary digital images by obtaining a matching metric between the two images, the apparatus including a filtering unit for obtaining $m \times n$ filtered images with respect to two arbitrary images by filtering the original image using predetermined filters, each having a unique combination of one of m scales and one of n orientations, where m and n are predetermined positive integers, and a matching metric calculating unit for calculating the matching metric defined by:

$$d_{m,n}(i, j) = \min_{1 \leq t \leq K} \left[\sum_{m,n} \left(\left| \frac{\mu_{m,n}^{(i)}}{\alpha(\mu_{m,n})} - \frac{\mu_{m,n \oplus t}^{(j)}}{\alpha(\mu_{m,n})} \right| \right) + \left(\left| \frac{\sigma_{m,n}^{(i)}}{\alpha(\sigma_{m,n})} - \frac{\sigma_{m,n \oplus t}^{(j)}}{\alpha(\sigma_{m,n})} \right| \right) \right] + b$$

with respect to the $m \times n$ filtered images, where the mean and variance of the pixel values of the respective images are μ and σ , the mean and variance of the pixel

values of the original image are μ_0 and σ_0 , K is a predetermined positive integer

representing the number of orientations to be considered, $b = \left\lfloor \frac{\mu_0^{(i)} - \mu_0^{(j)}}{\alpha(\mu_0)} \right\rfloor + \left\lfloor \frac{\sigma_0^{(i)} - \sigma_0^{(j)}}{\alpha(\sigma_0)} \right\rfloor$,

\oplus denotes a modulo shift function, and $\alpha(\cdot)$ is a predetermined scaling function.

To achieve the eleventh object of the present invention, there is provided a digital image texture analyzing method for evaluating the similarity of textures of two arbitrary digital images by obtaining a matching metric between the two images, the method including the steps of (a) obtaining filtered images using predetermined filters having different scale coefficients with respect to two arbitrary images, and (b) obtaining a matching metric by calculating the minimum value of the sum of absolute differences between the means of the pixel values with respect to an arbitrary filtered image and images filtered by filters having scale coefficients different from those of filters used for filtering the arbitrary image.

According to another aspect of the present invention, there is provided a digital image texture analyzing method for evaluating the similarity of textures of two arbitrary digital images by obtaining a matching metric between the two images, the method including the steps of (a) obtaining filtered images using predetermined filters having different scale coefficients with respect to two arbitrary images, and (b) obtaining a matching metric by calculating the minimum value of the sum of absolute differences between the variances of the pixel values with respect to an arbitrary filtered image and images filtered by filters having scale coefficients different from those of filters used for filtering the arbitrary image.

To achieve the twelfth object of the present invention, there is provided a computer-readable recording medium for storing program codes for performing a digital image texture analyzing method for evaluating the similarity of textures of two arbitrary digital images by obtaining a matching metric between the two images, the method including the steps of (a) obtaining $m \times n$ filtered images with respect to two arbitrary images by filtering the original image using predetermined filters, each having a unique combination of one of m scales and one of n orientations, where m and n are predetermined positive integers, and (b) calculating the matching metric defined by:

$$d(i, j) = \min_{\substack{p=0,1 \\ q=0,1}} \left[\sum_m^{S-1} \sum_n \left| \frac{\mu_{m+p,n}^i}{\alpha(\mu_{m+p,n})} - \frac{\mu_{m+q,n}^j}{\alpha(\mu_{m+q,n})} \right| + \left| \frac{\sigma_{m+p,n}^i}{\alpha(\sigma_{m+p,n})} - \frac{\sigma_{m+q,n}^j}{\alpha(\sigma_{m+q,n})} \right| \right]$$

with respect to the original image and the $m \times n$ filtered images, where the mean and variance of the pixel values of the respective images are μ and σ , S is a predetermined positive integer representing the number of scale coefficients to be considered, and $\alpha(\cdot)$ is a predetermined scaling function.

5 To achieve the thirteenth object of the present invention, there is provided a digital image texture analyzing apparatus for evaluating the similarity of textures of two arbitrary digital images by obtaining a matching metric between the two images, the apparatus including a filtering unit for obtaining $m \times n$ filtered images with respect to two arbitrary images by filtering the original image using
10 predetermined filters, each having a unique combination of one of m scales and one of n orientations, where m and n are predetermined positive integers, and a matching metric calculating unit for calculating the matching metric defined by:

$$d(i, j) = \min_{\substack{p=0,1 \\ q=0,1}} \left[\sum_m^{S-1} \sum_n \left| \frac{\mu_{m+p,n}^i}{\alpha(\mu_{m+p,n})} - \frac{\mu_{m+q,n}^j}{\alpha(\mu_{m+q,n})} \right| + \left| \frac{\sigma_{m+p,n}^i}{\alpha(\sigma_{m+p,n})} - \frac{\sigma_{m+q,n}^j}{\alpha(\sigma_{m+q,n})} \right| \right]$$

with respect to the original image and the $m \times n$ filtered images, where the mean and variance of the pixel values of the respective images are μ and σ , S is a
15 predetermined positive integer representing the number of scale coefficients to be considered, and $\alpha(\cdot)$ is a predetermined scaling function.

To achieve the fourteenth object of the present invention, there is provided a digital image texture analyzing method for evaluating the similarity of textures of two arbitrary digital images by obtaining a matching metric between the two images, the
20 method including the steps of (a) obtaining filtered images using predetermined filters having different orientation and scale coefficients with respect to two arbitrary images, and (b) obtaining a matching metric by calculating the minimum value of the sum of absolute differences between the variances and means of the pixel values with respect to an arbitrary filtered image and images filtered by filters having scale

and orientation coefficients different from those of filters used for filtering the arbitrary image.

BRIEF DESCRIPTION OF THE DRAWINGS

5 The above objects and advantages of the present invention will become more apparent by describing in detail a preferred embodiment thereof with reference to the attached drawings in which:

FIG. 1 illustrates a conventional digital image texture analyzing method;

FIG. 2 is block diagram illustrating a digital image texture analyzing apparatus
10 according to an embodiment of the present invention;

FIG. 3 is a flow chart illustrating major steps of a digital image texture
analyzing method according to an embodiment of the present invention;

FIG. 4 is block diagram illustrating a digital image texture analyzing apparatus
according to another embodiment of the present invention; and

15 FIGS. 5 through 7 are flow charts illustrating major steps of a digital image texture analyzing method according to another embodiments of the present invention performed by the apparatus shown in FIG. 4.

DESCRIPTION OF THE PREFERRED EMBODIMENTS

20 Preferred embodiments of the present invention will now be described in detail with reference to the accompanying drawings.

FIG. 2 is block diagram illustrating a digital image texture analyzing apparatus according to an embodiment of the present invention, and FIG. 3 is a flow chart illustrating major steps of a digital image texture analyzing method according to an
25 embodiment of the present invention. FIG. 3 will be occasionally referred to below.

First, referring to FIG. 2, the digital image texture analyzing apparatus according to an embodiment of the present invention includes a first mean/variance calculating unit 202, a filtering unit 204, a second mean/variance calculating unit 206 and a texture descriptor setting unit 208.

30 The operation of the digital image texture analyzing apparatus will now be described.

The first mean/variance calculating unit 202 calculates the mean and variance of the pixel values of an original image, μ_0 and σ_0 . (step 302). Next, the filtering unit 204 obtains $m \times n$ filtered images using Garbor filters, each having a unique

combination of one of m scales and one of n orientations (m and n are predetermined positive integers) (step 304). Then, the second mean/variance calculating unit 206 calculates the means and variances of the respective filtered images $\mu_{11}, \sigma_{11}, \mu_{21}, \sigma_{21}, \dots, \mu_{mn}, \sigma_{mn}$ (step 306).

Now, the texture descriptor setting unit 208 obtains a texture descriptor in the form of $\{ \mu_0, \sigma_0, \mu_{11}, \sigma_{11}, \mu_{21}, \sigma_{21}, \dots, \mu_{mn}, \sigma_{mn} \}$, containing the mean and variance of the pixel values of the original image obtained in the step 302, μ_0 and σ_0 , and the means and variances of the respective filtered images $\mu_{11}, \sigma_{11}, \mu_{21}, \sigma_{21}, \dots, \mu_{mn}, \sigma_{mn}$ obtained in the step 306 (step 308).

The texture descriptor is used in indexing images. The indexed images form an image database. If the image database is formed, of the two indexed images, an image having a texture descriptor similar to a texture descriptor of a query image can be searched for. In particular, in order to search for an image similar to a query image using a texture descriptor having the mean and variance of the pixel values of an original image as texture features based on the analyzing method of the present invention, simulation was performed. The simulation result showed that the searching efficiency was improved.

FIG. 4 is a block diagram of a digital image texture analyzing apparatus according to another embodiment of the present invention, and FIG. 5 is a flow chart illustrating major steps of a digital image texture analyzing method according to another embodiment of the present invention. FIG. 5 will be occasionally referred to below.

Referring to FIG. 4, the digital image texture analyzing apparatus according to another embodiment of the present invention includes a filtering unit 402 and a matching metric calculating unit 404.

The operation of the digital image texture analyzing apparatus will now be described.

The filtering unit 402 obtains $m \times n$ filtered images with respect to two arbitrary images using Garbor filters, each having a unique combination of one of m scales and one of n orientations (m and n are predetermined positive integers) (step 502).

The matching metric calculating unit 404 calculates the matching metric defined by:

$$d(i, j) = \sum_{m,n} d_{m,n}(i, j) + b$$

with respect to the original image and the $m \times n$ filtered images, where the mean and variance of the pixel values of the respective images are μ and σ , the mean and variance of the pixel values of the original image are μ_0 and σ_0 ,

$$b = \left| \frac{\mu_0^{(i)} - \mu_0^{(j)}}{\alpha(\mu_0)} \right| + \left| \frac{\sigma_0^{(i)} - \sigma_0^{(j)}}{\alpha(\sigma_0)} \right|, \text{ and } d_{m,n}(i, j) = \left| \frac{\mu_{m,n}^{(i)} - \mu_{m,n}^{(j)}}{\alpha(\mu_{m,n})} \right| + \left| \frac{\sigma_{m,n}^{(i)} - \sigma_{m,n}^{(j)}}{\alpha(\sigma_{m,n})} \right| \text{ (step 504). Here,}$$

5 $\alpha(\cdot)$ denotes a scaling function for normalizing individual feature components. Thus, $\alpha(\mu_{m,n})$ and $\alpha(\sigma_{m,n})$ are scaled mean and variance of the respective feature components of the descriptors over an entire database. The calculated matching metric indicates a difference between texture features of the two images. In other words, if the value of the calculated matching metric is smaller, the textures of the two images are determined to be similar to each other.

10 According to the above-described digital image texture analyzing method, the similarity between two arbitrary digital images can be determined by obtaining the matching metric between two images. Also, the digital image texture analyzing method can be applied to search for an image using the texture features of the image.

15 FIG. 6 is a flow chart illustrating major steps of a digital image texture analyzing method according to another embodiments of the present invention performed by the apparatus shown in FIG. 4.

20 In this embodiment, the filtering unit 402 obtains $m \times n$ filtered images with respect to two arbitrary images by filtering the original image using predetermined filters, each having a unique combination of one of m scales and one of n orientations (m and n are predetermined positive integers) (step 602).

Next, the matching metric calculating unit 404 calculates the matching metric defined by:

$$d_{m,n}(i, j) = \min_{1 \leq t \leq K} \left[\sum_{m,n} \left(\left| \frac{\mu_{m,n}^{(i)}}{\alpha(\mu_{m,n})} - \frac{\mu_{m,n \oplus t}^{(j)}}{\alpha(\mu_{m,n})} \right| + \left| \frac{\sigma_{m,n}^{(i)}}{\alpha(\sigma_{m,n})} - \frac{\sigma_{m,n \oplus t}^{(j)}}{\alpha(\sigma_{m,n})} \right| \right) \right] + b$$

with respect to the $m \times n$ filtered images, where the mean and variance of the pixel values of the respective images are μ and σ , the mean and variance of the pixel values of the original image are μ_0 and σ_0 , K is a predetermined positive integer representing the number of orientations to be considered,

$$b = \left| \frac{\mu_0^{(i)} - \mu_0^{(j)}}{\alpha(\mu_0)} \right| + \left| \frac{\sigma_0^{(i)} - \sigma_0^{(j)}}{\alpha(\sigma_0)} \right|, \oplus \text{ denotes a modulo shift function, and } \alpha(\cdot) \text{ is a}$$

predetermined scaling function (step 604). The calculated matching metric indicates a difference between texture features of the two images. In other words, if the value of the calculated matching metric is smaller, the textures of the two images are determined to be similar to each other. In particular, only when the texture features of two images are similar and one image is rotated relative to the other image, the thus-calculated matching metric allows accurate analysis, that is, it is determined that the two images have similar texture features. In other words, even when one image is just rotated relative to the other image, it is determined that the texture features of the two images are similar to each other, unlike in the conventional analyzing method. According to the above-described digital image texture analyzing method, the similarity between two arbitrary digital images can be determined by obtaining the matching metric between two images. Also, the digital image texture analyzing method can be applied to search for an image using the texture features of the image.

In the above-described embodiment, it has been described that the matching metric is obtained by calculating the sum of absolute differences in the mean of the pixel values and the sum of absolute differences between the variances of the pixel values, and calculating the minimum value of an added value of the calculated sums, with respect to an arbitrary filtered image and images filtered by Garbor filters having orientation coefficients different from those of filters used for filtering the arbitrary image.

However, the matching metric can also be obtained by calculating the minimum value of the sum of absolute differences in the mean of the pixel values with respect to an arbitrary filtered image and images filtered by Garbor filters having orientation coefficients different from those of filters used for filtering the arbitrary image. Alternatively, the matching metric can also be obtained by calculating the minimum value of the sum of absolute differences between the variances of the pixel

values with respect to an arbitrary filtered image and images filtered by Garbor filters having orientation coefficients different from those of filters used for filtering the arbitrary image.

FIG. 7 is a flow chart illustrating major steps of a digital image texture analyzing method according to another embodiments of the present invention performed by the apparatus shown in FIG. 4.

In this embodiment, the filtering unit 402 obtains $m \times n$ filtered images with respect to two arbitrary images using Garbor filters, each having a unique combination of one of m scales and one of n orientations (m and n are positive integers) (step 702).

Next, the matching metric calculating unit 404 calculates the matching metric defined by:

$$d(i, j) = \min_{p=0,1} \sum_{q=0,1}^{S-1} \sum_m \sum_n \left| \frac{\mu_{m+p,n}^i}{\alpha(\mu_{m+p,n})} - \frac{\mu_{m+q,n}^j}{\alpha(\mu_{m+q,n})} \right| + \left| \frac{\sigma_{m+p,n}^i}{\alpha(\sigma_{m+p,n})} - \frac{\sigma_{m+q,n}^j}{\alpha(\sigma_{m+q,n})} \right| + b$$

with respect to the original image and the $m \times n$ filtered images, where the mean and variance of the pixel values of the respective images are μ and σ , the mean and variance of the pixel values of the original image are μ_0 and σ_0 ,

$$b = \left| \frac{\mu_0^{(i)} - \mu_0^{(j)}}{\alpha(\mu_0)} \right| + \left| \frac{\sigma_0^{(i)} - \sigma_0^{(j)}}{\alpha(\sigma_0)} \right|, S \text{ is a predetermined positive integer representing the}$$

number of scale coefficients to be considered, and $\alpha(\)$ is a predetermined scaling function (step 704). The calculated matching metric indicates a difference between texture features of the two images. In other words, if the value of the calculated matching metric is smaller, the textures of the two images are determined to be similar to each other. In particular, only when the texture features of two images are similar and one image is enlarged or reduced relative to the other image, the thus-calculated matching metric allows accurate analysis, that is, it is determined that the two images have similar texture features. In other words, even when one image is just enlarged or reduced relative to the other image, it is determined that the texture features of the two images are similar to each other, unlike in the conventional analyzing method. According to the above-described digital image texture analyzing method, the similarity between two arbitrary digital images can be determined by

obtaining the matching metric between two digital images. Also, the digital image texture analyzing method can be applied to search for an image using the texture features of the image.

In the above-described embodiment, it has been described that the matching metric is obtained by calculating the sum of absolute differences in the mean of the pixel values and the sum of absolute differences between the variances of the pixel values, and calculating the minimum value of an added value of the calculated sums, with respect to an arbitrary filtered image and images filtered by Garbor filters having scale coefficients different from those of filters used for filtering the arbitrary image.

The matching metric can also be obtained by calculating the minimum value of the sum of absolute differences in the mean of the pixel values with respect to an arbitrary filtered image and images filtered by Garbor filters having scale coefficients different from those of filters used for filtering the arbitrary image. Alternatively, the matching metric can also be obtained by calculating the minimum value of the sum of absolute differences between the variances of the pixel values with respect to an arbitrary filtered image and images filtered by Garbor filters having orientation coefficients different from those of filters used for filtering the arbitrary image.

The technologies applied to the embodiments described above with reference to FIGS. 6 and 7 can be used in appropriate combinations as understood by one skilled in the art. According to the digital image texture analyzing method of the present invention, even when one image is rotated, enlarged or reduced relative to the other image, the two images can be accurately analyzed, that is, it is determined that they have similar texture features. The embodiments based on the combinations of the above-described embodiments shown in FIGS. 6 and 7 will not be described herein.

As described above, in the digital image texture analyzing method according to the present invention, the texture features of images can be analyzed or compared accurately. Also, according to the digital image texture analyzing method of the present invention, even when one image is rotated, enlarged or reduced relative to the other image, accurate analysis of the two images is allowed, that is, it is determined that they have similar texture features.

What is claimed is:

1 1. A digital image texture analyzing method comprising the step of
2 obtaining a texture descriptor including a mean of pixel values of an original image
3 as a texture feature.

1 2. The method according to claim 1, wherein the texture feature further
2 includes a variance of the pixel values of the original image.

1 3. A digital image texture analyzing method comprising the step of
2 obtaining a texture descriptor including a variance of pixel values of an original
3 image as a texture feature.

1 4. A digital image texture analyzing method comprising the steps of:
2 (a) obtaining a mean (μ_0) and a variance (σ_0) of pixel values of an original
3 image;
4 (b) obtaining $m \times n$ filtered images by filtering the original image using
5 predetermined filters, each having a unique combination of one of m scales and one
6 of n orientations, where m and n are predetermined positive integers;
7 (c) calculating means ($\mu_{11}, \mu_{21}, \dots, \mu_{mn}$) and variances ($\sigma_{11}, \sigma_{21}, \dots, \sigma_{mn}$) of
8 the respective filtered images; and
9 (d) obtaining a texture descriptor having the mean (μ_0) and the variance (σ_0)
10 of the pixel values of the original image obtained in the step (a), and the means ($\mu_{11},$
11 $\mu_{21}, \dots, \mu_{mn}$) and the variances ($\sigma_{11}, \sigma_{21}, \dots, \sigma_{mn}$) of the respective filtered images
12 obtained in the step (c), as texture features.

1 5. The method according to claim 4, wherein the predetermined filters are
2 Garbor filters.

1 6. A computer-readable recording medium for storing program codes for
2 performing a digital image texture analyzing method comprising the steps of:
3 (a) obtaining a mean (μ_0) and a variance (σ_0) of the pixel values of an
4 original image;

(b) obtaining $m \times n$ filtered images by filtering the original image using predetermined filters, each having a unique combination of one of m scales and one of n orientations, where m and n are predetermined positive integers;

(c) calculating means ($\mu_{11}, \mu_{21}, \dots, \mu_{mn}$) and variances ($\sigma_{11}, \sigma_{21}, \dots, \sigma_{mn}$) of the respective filtered images; and

(d) obtaining a texture descriptor having the mean (μ_0) and the variance (σ_0) of the pixel values of the original image obtained in the step (a), and the means ($\mu_{11}, \mu_{21}, \dots, \mu_{mn}$) and the variances ($\sigma_{11}, \sigma_{21}, \dots, \sigma_{mn}$) of the respective filtered images, obtained in the step (c), as texture features.

7. The recording medium according to claim 6, wherein the predetermined filters are Garbor filters.

8. A digital image texture analyzing apparatus comprising:

a mean/variance calculating unit which calculates a mean and a variance of pixel values of an original image; and

a texture descriptor setting unit which sets the mean and the variance as texture descriptors.

9. The apparatus according to claim 8, further comprising a filtering unit which filters the original image using predetermined filters, each having a unique combination of one of m scales and one of n orientations, where m and n are predetermined positive integers, to obtain $m \times n$ filtered images,

wherein the mean/variance calculating unit obtains the mean (μ_0) and the variance (σ_0) of the pixel values of the original image and the means ($\mu_{11}, \mu_{21}, \dots, \mu_{mn}$) and the variances ($\sigma_{11}, \sigma_{21}, \dots, \sigma_{mn}$) of the respective filtered images, and

wherein the texture descriptor setting unit obtains a texture descriptor having the mean (μ_0) and the variance (σ_0) of the pixel values of the original image, and the means ($\mu_{11}, \mu_{21}, \dots, \mu_{mn}$) and the variances ($\sigma_{11}, \sigma_{21}, \dots, \sigma_{mn}$) of the respective filtered images, as texture features.

1 10. A digital image searching method comprising searching for an image
2 having a similar texture descriptor to a query image using a texture descriptor having
3 a mean and a variance of the pixel values of an original image as texture features.

1 11. A digital image texture analyzing method for evaluating a similarity of
2 textures of two arbitrary digital images by obtaining a matching metric between the
3 two arbitrary images, the method comprising calculating a matching metric including
4 absolute differences between means of pixel values of the two arbitrary digital
5 images.

1 12. The method according to claim 11, wherein the matching metric further
2 includes absolute differences between variances of the pixel values of the original
3 image.

1 13. A digital image texture analyzing method for evaluating a similarity of
2 textures of two arbitrary digital images by obtaining a matching metric between the
3 two arbitrary images, the method comprising calculating a matching metric including
4 absolute differences between variances of pixel values of the two arbitrary digital
5 images.

1 14. A digital image texture analyzing method for evaluating a similarity of
2 textures of two arbitrary digital images by obtaining a matching metric between the
3 two arbitrary images, the method comprising the steps of:

4 (a) obtaining $m \times n$ filtered images with respect to the two arbitrary images
5 using predetermined filters, each having a unique combination of one of m scales
6 and one of n orientations, where m and n are predetermined positive integers; and

7 (b) calculating the matching metric defined by:

$$d(i, j) = \sum_{m, n} d_{m, n}(i, j) + b$$

with respect to an original image and the $m \times n$ filtered images, where a mean and a variance of pixel values of the respective filtered images are μ and σ , a mean and a variance of pixel values of the original image are μ_0 and σ_0 ,

$$b = \left| \frac{\mu_0^{(i)} - \mu_0^{(j)}}{\alpha(\mu_0)} \right| + \left| \frac{\sigma_0^{(i)} - \sigma_0^{(j)}}{\alpha(\sigma_0)} \right|, \quad d_{m,n}(i, j) = \left| \frac{\mu_{m,n}^{(i)} - \mu_{m,n}^{(j)}}{\alpha(\mu_{m,n})} \right| + \left| \frac{\sigma_{m,n}^{(i)} - \sigma_{m,n}^{(j)}}{\alpha(\sigma_{m,n})} \right|, \text{ and } \alpha(\cdot) \text{ is a}$$

predetermined scaling function.

15. The method according to claim 14, wherein the predetermined filters are Garbor filters.

16. A computer-readable recording medium for storing program codes for performing a digital image texture analyzing method for evaluating a similarity of textures of two arbitrary digital images by obtaining a matching metric between the two arbitrary images, the method comprising the steps of:

- (a) obtaining $m \times n$ filtered images with respect to the two arbitrary images using predetermined filters, each having a unique combination of one of m scales and one of n orientations, where m and n are predetermined positive integers; and
- (b) calculating the matching metric defined by:

$$d(i, j) = \sum_{m, n} d_{m, n}(i, j) + b$$

with respect to an original image and the $m \times n$ filtered images, where a mean and a variance of pixel values of the respective filtered images are μ and σ , a mean and a variance of pixel values of the original image are μ_0 and σ_0 ,

$$b = \left| \frac{\mu_0^{(i)} - \mu_0^{(j)}}{\alpha(\mu_0)} \right| + \left| \frac{\sigma_0^{(i)} - \sigma_0^{(j)}}{\alpha(\sigma_0)} \right|, \quad d_{m,n}(i, j) = \left| \frac{\mu_{m,n}^{(i)} - \mu_{m,n}^{(j)}}{\alpha(\mu_{m,n})} \right| + \left| \frac{\sigma_{m,n}^{(i)} - \sigma_{m,n}^{(j)}}{\alpha(\sigma_{m,n})} \right|, \text{ and } \alpha(\cdot) \text{ is a}$$

predetermined scaling function.

17. The recording medium according to claim 16, wherein the predetermined filters are Garbor filters.

18. A digital image texture analyzing apparatus for evaluating a similarity of textures of two arbitrary digital images by obtaining a matching metric between the two arbitrary images, the apparatus comprising:

(a) a filtering unit which obtains $m \times n$ filtered images with respect to the two arbitrary images using predetermined filters, each having a unique combination of one of m scales and one of n orientations, where m and n are predetermined positive integers; and

(b) a matching metric calculating unit which calculates the matching metric defined by:

$$d(i, j) = \sum_{m, n} d_{m, n}(i, j) + b$$

with respect to an original image and the $m \times n$ filtered images, where a mean and a variance of pixel values of the respective filtered images are μ and σ , a mean and a variance of pixel values of the original image are μ_0 and σ_0 ,

$$b = \left| \frac{\mu_0^{(i)} - \mu_0^{(j)}}{\alpha(\mu_0)} \right| + \left| \frac{\sigma_0^{(i)} - \sigma_0^{(j)}}{\alpha(\sigma_0)} \right|, \quad d_{m, n}(i, j) = \left| \frac{\mu_{m, n}^{(i)} - \mu_{m, n}^{(j)}}{\alpha(\mu_{m, n})} \right| + \left| \frac{\sigma_{m, n}^{(i)} - \sigma_{m, n}^{(j)}}{\alpha(\sigma_{m, n})} \right|, \text{ and } \alpha(\cdot) \text{ is a}$$

predetermined scaling function.

19. The apparatus according to claim 18, wherein the predetermined filters are Garbor filters.

20. A digital image texture analyzing method for evaluating a similarity of textures of two arbitrary digital images by obtaining a matching metric between the two arbitrary images, the method comprising the steps of:

(a) obtaining filtered images using predetermined filters having different orientation coefficients with respect to the two arbitrary images; and

(b) obtaining a matching metric by calculating a minimum value of a sum of absolute differences between means of pixel values with respect to an arbitrary filtered image and images filtered by filters having orientation coefficients different from those of filters used for filtering the arbitrary image.

1 21. The method according to claim 20, wherein the predetermined filters
2 are Garbor filters.

1 22. A digital image texture analyzing method for evaluating a similarity of
2 textures of two arbitrary digital images by obtaining a matching metric between the
3 two arbitrary images, the method comprising the steps of:

4 (a) obtaining filtered images using predetermined filters having different
5 orientation coefficients with respect to the two arbitrary images; and

6 (b) obtaining a matching metric by calculating a minimum value of a sum of
7 absolute differences between variances of pixel values with respect to an arbitrary
8 filtered image and images filtered by filters having orientation coefficients different
9 from those of filters used for filtering the arbitrary image.

1 23. The method according to claim 22, wherein the predetermined filters
2 are Garbor filters.

1 24. The method according to claim 22, wherein the step (b) includes the
2 step of obtaining a matching metric by calculating the minimum value of an added
3 value of the sums of absolute differences between the means and the variance of
4 the pixel values with respect to the arbitrary filtered image and the images filtered by
5 the filters having orientation coefficients different from those of the filters used for
6 filtering the arbitrary image.

1 25. The method according to claim 22, wherein the step (a) includes the
2 step of (a') obtaining $m \times n$ filtered images with respect to the two arbitrary images by
3 filtering an original image using predetermined filters, each having a unique
4 combination of one of m scales and one of n orientations, where m and n are
5 predetermined positive integers; and wherein the step (b) includes the step of (b')
6 calculating the matching metric defined by:

$$d_{m,n}(i, j) = \min_{1 \leq t \leq K} \left[\sum_{m,n} \left(\left| \frac{\mu_{m,n}^{(i)}}{\alpha(\mu_{m,n})} - \frac{\mu_{m,n \oplus t}^{(j)}}{\alpha(\mu_{m,n})} \right| \right) + \left(\left| \frac{\sigma_{m,n}^{(i)}}{\alpha(\sigma_{m,n})} - \frac{\sigma_{m,n \oplus t}^{(j)}}{\alpha(\sigma_{m,n})} \right| \right) \right]$$

with respect to the $m \times n$ filtered images, where a mean and a variance of pixel values of the respective images are μ and σ , K is a predetermined positive integer representing a number of orientations to be considered, \oplus denotes a modulo shift function, and $\alpha(\cdot)$ is a predetermined scaling function.

26. The method according to claim 25, wherein the step of (b') calculating the matching metric further defined by:

$$d_{m,n}(i, j) = \min_{1 \leq t \leq K} \left[\sum_{m,n} \left(\left| \frac{\mu_{m,n}^{(i)}}{\alpha(\mu_{m,n})} - \frac{\mu_{m,n \oplus t}^{(j)}}{\alpha(\mu_{m,n})} \right| \right) + \left(\left| \frac{\sigma_{m,n}^{(i)}}{\alpha(\sigma_{m,n})} - \frac{\sigma_{m,n \oplus t}^{(j)}}{\alpha(\sigma_{m,n})} \right| \right) \right] + b$$

with respect to the $m \times n$ filtered images, where the mean and the variance of the pixel values of the respective images are μ and σ , a mean and a variance of pixel values of the original image are μ_0 and σ_0 , K is the predetermined positive integer representing the number of orientations to be considered, $b = \left| \frac{\mu_0^{(i)} - \mu_0^{(j)}}{\alpha(\mu_0)} \right| + \left| \frac{\sigma_0^{(i)} - \sigma_0^{(j)}}{\alpha(\sigma_0)} \right|$, \oplus denotes the modulo shift function, and $\alpha(\cdot)$ is the predetermined scaling function.

27. A computer-readable recording medium for storing program codes for performing a digital image texture analyzing method for evaluating a similarity of textures of two arbitrary digital images by obtaining a matching metric between the two arbitrary images, the method comprising the steps of:

- (a) obtaining $m \times n$ filtered images with respect to the two arbitrary images using predetermined filters, each having a unique combination of one of m scales and one of n orientations, where m and n are predetermined positive integers; and
- (b) calculating the matching metric defined by:

$$d_{m,n}(i, j) = \min_{1 \leq t \leq K} \left[\sum_{m,n} \left(\left| \frac{\mu_{m,n}^{(i)}}{\alpha(\mu_{m,n})} - \frac{\mu_{m,n \oplus t}^{(j)}}{\alpha(\mu_{m,n})} \right| \right) + \left(\left| \frac{\sigma_{m,n}^{(i)}}{\alpha(\sigma_{m,n})} - \frac{\sigma_{m,n \oplus t}^{(j)}}{\alpha(\sigma_{m,n})} \right| \right) \right]$$

with respect to the $m \times n$ filtered images, where a mean and a variance of the pixel values of the respective images are μ and σ , K is a predetermined positive integer

11 representing a number of orientations to be considered, \oplus denotes a modulo shift
 12 function, and $\alpha(\cdot)$ is a predetermined scaling function.

1 28. The computer-readable recording medium according to claim 27,
 2 wherein the step of (b') calculating the matching metric further defined by:

$$d_{m,n}(i, j) = \min_{1 \leq t \leq K} \left[\sum_{m,n} \left(\left| \frac{\mu_{m,n}^{(i)}}{\alpha(\mu_{m,n})} - \frac{\mu_{m,n \oplus t}^{(j)}}{\alpha(\mu_{m,n})} \right| \right) + \left(\left| \frac{\sigma_{m,n}^{(i)}}{\alpha(\sigma_{m,n})} - \frac{\sigma_{m,n \oplus t}^{(j)}}{\alpha(\sigma_{m,n})} \right| \right) \right] + b$$

3 with respect to the $m \times n$ filtered images, where the mean and the variance of the
 4 pixel values of the respective images are μ and σ , a mean and a variance of pixel
 5 values of the original image are μ_0 and σ_0 , K is the predetermined positive integer
 6 representing the number of orientations to be considered,

7 $b = \left| \frac{\mu_0^{(i)} - \mu_0^{(j)}}{\alpha(\mu_0)} \right| + \left| \frac{\sigma_0^{(i)} - \sigma_0^{(j)}}{\alpha(\sigma_0)} \right|$, \oplus denotes the modulo shift function, and $\alpha(\cdot)$ is the
 8 predetermined scaling function.

1 29. A digital image texture analyzing apparatus for evaluating a similarity of
 2 textures of two arbitrary digital images by obtaining a matching metric between the
 3 two arbitrary digital images, the apparatus comprising:

4 a filtering unit which obtains $m \times n$ filtered images with respect to the two
 5 arbitrary images by filtering an original image using predetermined filters, each
 6 having a unique combination of one of m scales and one of n orientations, where m
 7 and n are predetermined positive integers; and

8 a matching metric calculating unit which calculates the matching metric
 9 defined by:

$$d_{m,n}(i, j) = \min_{1 \leq t \leq K} \left[\sum_{m,n} \left(\left| \frac{\mu_{m,n}^{(i)}}{\alpha(\mu_{m,n})} - \frac{\mu_{m,n \oplus t}^{(j)}}{\alpha(\mu_{m,n})} \right| \right) + \left(\left| \frac{\sigma_{m,n}^{(i)}}{\alpha(\sigma_{m,n})} - \frac{\sigma_{m,n \oplus t}^{(j)}}{\alpha(\sigma_{m,n})} \right| \right) \right] + b$$

10 with respect to the $m \times n$ filtered images, where a mean and a variance of pixel
 11 values of the respective filtered images are μ and σ , a mean and a variance of

12 pixel values of the original image are μ_0 and σ_0 , K is a predetermined positive
 13 integer representing the number of orientations to be considered,
 14 $b = \left| \frac{\mu_0^{(i)} - \mu_0^{(j)}}{\alpha(\mu_0)} \right| + \left| \frac{\sigma_0^{(i)} - \sigma_0^{(j)}}{\alpha(\sigma_0)} \right|$, \oplus denotes a modulo shift function, and $\alpha(\)$ is a
 15 predetermined scaling function.

1 30. A digital image texture analyzing method for evaluating a similarity of
 2 textures of two arbitrary digital images by obtaining a matching metric between the
 3 two arbitrary images, the method comprising the steps of:

4 (a) obtaining filtered images using predetermined filters having different scale
 5 coefficients with respect to the two arbitrary images; and

6 (b) obtaining a matching metric by calculating a minimum value of a sum of
 7 absolute differences between means of pixel values with respect to an arbitrary
 8 filtered image and images filtered by filters having scale coefficients different from
 9 those of filters used for filtering the arbitrary image.

1 31. The method according to claim 30, wherein the predetermined filters
 2 are Garbor filters.

1 32. A digital image texture analyzing method for evaluating a similarity of
 2 textures of two arbitrary digital images by obtaining a matching metric between the
 3 two arbitrary images, the method comprising the steps of:

4 (a) obtaining filtered images using predetermined filters having different scale
 5 coefficients with respect to the two arbitrary images; and

6 (b) obtaining a matching metric by calculating a minimum value of a sum of
 7 absolute differences between variances of pixel values with respect to an arbitrary
 8 filtered image and images filtered by filters having scale coefficients different from
 9 those of filters used for filtering the arbitrary image.

1 33. The method according to claim 32, wherein the predetermined filters
 2 are Garbor filters.

34. The method according to claim 32, wherein the step (b) includes the step of (b') obtaining a matching metric by calculating the minimum value of an added value of the sums of absolute differences between the means and a variance of the pixel values with respect to an arbitrary filtered image and images filtered by filters having scale filters different from those of filters used for filtering the arbitrary image.

35. The method according to claim 32, wherein the step (a) includes the step of (a') obtaining $m \times n$ filtered images with respect to the two arbitrary images by filtering an original image using predetermined filters, each having a unique combination of one of m scales and one of n orientations, where m and n are predetermined positive integers; and wherein the step (b) includes the step of (b') calculating the matching metric defined by:

$$d(i, j) = \min_{p=0,1} \sum_{q=0,1} \left[\sum_m \sum_n \left| \frac{\mu_{m+p,n}^i}{\alpha(\mu_{m+p,n})} - \frac{\mu_{m+q,n}^j}{\alpha(\mu_{m+q,n})} \right| + \left| \frac{\sigma_{m+p,n}^i}{\alpha(\sigma_{m+p,n})} - \frac{\sigma_{m+q,n}^j}{\alpha(\sigma_{m+q,n})} \right| \right]$$

with respect to the original image and the $m \times n$ filtered images, where a mean and a variance of pixel values of the respective images are μ and σ , S is a predetermined positive integer representing a number of scale coefficients to be considered, and $\alpha(\)$ is a predetermined scaling function.

36. The method according to claim 35, wherein the step (b') includes the step of calculating the matching metric defined by:

$$d(i, j) = \min_{p=0,1} \sum_{q=0,1} \left[\sum_m \sum_n \left| \frac{\mu_{m+p,n}^i}{\alpha(\mu_{m+p,n})} - \frac{\mu_{m+q,n}^j}{\alpha(\mu_{m+q,n})} \right| + \left| \frac{\sigma_{m+p,n}^i}{\alpha(\sigma_{m+p,n})} - \frac{\sigma_{m+q,n}^j}{\alpha(\sigma_{m+q,n})} \right| \right] + b$$

with respect to the original image and the $m \times n$ filtered images, where the mean and the variance of the pixel values of the respective images are μ and σ , a mean and a variance of pixel values of the original image are μ_0 and σ_0 ,

6 $b = \left| \frac{\mu_0^{(i)} - \mu_0^{(j)}}{\alpha(\mu_0)} \right| + \left| \frac{\sigma_0^{(i)} - \sigma_0^{(j)}}{\alpha(\sigma_0)} \right|$, S is the predetermined positive integer representing the
7 number of scale coefficients to be considered, and $\alpha(\cdot)$ is the predetermined scaling
8 function.

1 37. A computer-readable recording medium for storing program codes for
2 performing a digital image texture analyzing method for evaluating a similarity of
3 textures of two arbitrary digital images by obtaining a matching metric between the
4 two arbitrary images, the method comprising the steps of:

5 (a) obtaining $m \times n$ filtered images with respect to the two arbitrary images by
6 filtering an original image using predetermined filters, each having a unique
7 combination of one of m scales and one of n orientations, where m and n are
8 predetermined positive integers; and

9 (b) calculating the matching metric defined by:

$$d(i, j) = \min_{\substack{p=0,1 \\ q=0,1}} \left[\sum_{m=0}^{S-1} \sum_n \left| \frac{\mu_{m+p,n}^i}{\alpha(\mu_{m+p,n})} - \frac{\mu_{m+q,n}^j}{\alpha(\mu_{m+q,n})} \right| + \left| \frac{\sigma_{m+p,n}^i}{\alpha(\sigma_{m+p,n})} - \frac{\sigma_{m+q,n}^j}{\alpha(\sigma_{m+q,n})} \right| \right]$$

10 with respect to the original image and the $m \times n$ filtered images, where a mean and a
11 variance of pixel values of the respective images are μ and σ , S is a predetermined
12 positive integer representing a number of scale coefficients to be considered, and
13 $\alpha(\cdot)$ is a predetermined scaling function.

1 38. The computer-readable recording medium according to claim 37,
2 wherein the step (b) includes the step of calculating the matching metric defined by:

$$d(i, j) = \min_{\substack{p=0,1 \\ q=0,1}} \left[\sum_{m=0}^{S-1} \sum_n \left| \frac{\mu_{m+p,n}^i}{\alpha(\mu_{m+p,n})} - \frac{\mu_{m+q,n}^j}{\alpha(\mu_{m+q,n})} \right| + \left| \frac{\sigma_{m+p,n}^i}{\alpha(\sigma_{m+p,n})} - \frac{\sigma_{m+q,n}^j}{\alpha(\sigma_{m+q,n})} \right| \right] + b$$

3 with respect to the original image and the $m \times n$ filtered images, where the mean and
4 the variance of the pixel values of the respective images are μ and σ , a mean and
5 a variance of pixel values of the original image are μ_0 and σ_0 ,

6 $b = \left| \frac{\mu_0^{(i)} - \mu_0^{(j)}}{\alpha(\mu_0)} \right| + \left| \frac{\sigma_0^{(i)} - \sigma_0^{(j)}}{\alpha(\sigma_0)} \right|$, S is the predetermined positive integer representing the
7 number of scale coefficients to be considered, and $\alpha(\)$ is the predetermined scaling
8 function.

1 39. A digital image texture analyzing apparatus for evaluating a similarity of
2 textures of two arbitrary digital images by obtaining a matching metric between the
3 two arbitrary images, the apparatus comprising:

4 a filtering unit which obtains $m \times n$ filtered images with respect to the two
5 arbitrary images by filtering an original image using predetermined filters, each
6 having a unique combination of one of m scales and one of n orientations, where m
7 and n are predetermined positive integers; and

8 a matching metric calculating unit which calculates the matching metric
9 defined by:

$$d(i, j) = \min_{\substack{p=0,1 \\ q=0,1}} \left[\sum_m^{S-1} \sum_n \left| \frac{\mu_{m+p,n}^i}{\alpha(\mu_{m+p,n})} - \frac{\mu_{m+q,n}^j}{\alpha(\mu_{m+q,n})} \right| + \left| \frac{\sigma_{m+p,n}^i}{\alpha(\sigma_{m+p,n})} - \frac{\sigma_{m+q,n}^j}{\alpha(\sigma_{m+q,n})} \right| \right]$$

10 with respect to the original image and the $m \times n$ filtered images, where a mean and a
11 variance of pixel values of the respective images are μ and σ , S is a predetermined
12 positive integer representing a number of scale coefficients to be considered, and
13 $\alpha(\)$ is a predetermined scaling function.

1 40. The apparatus according to claim 39, wherein the predetermined filters
2 are Garbor filters.

1 41. The apparatus according to claim 39, wherein the matching metric
2 calculating unit calculates the matching metric defined by:

$$d(i, j) = \min_{\substack{p=0,1 \\ q=0,1}} \left[\sum_m^{S-1} \sum_n \left| \frac{\mu_{m+p,n}^i}{\alpha(\mu_{m+p,n})} - \frac{\mu_{m+q,n}^j}{\alpha(\mu_{m+q,n})} \right| + \left| \frac{\sigma_{m+p,n}^i}{\alpha(\sigma_{m+p,n})} - \frac{\sigma_{m+q,n}^j}{\alpha(\sigma_{m+q,n})} \right| \right] + b$$

3 with respect to the original image and the $m \times n$ filtered images, where the mean and
4 the variance of the pixel values of the respective images are μ and σ , a mean and
5 a variance of pixel values of the original image are μ_0 and σ_0 ,

6
$$b = \left| \frac{\mu_0^{(i)} - \mu_0^{(j)}}{\alpha(\mu_0)} \right| + \left| \frac{\sigma_0^{(i)} - \sigma_0^{(j)}}{\alpha(\sigma_0)} \right|$$
, S is the predetermined positive integer representing the
7 number of scale coefficients to be considered, and $\alpha(\)$ is the predetermined scaling
8 function.

1 42. A digital image texture analyzing method for evaluating a similarity of
2 textures of two arbitrary digital images by obtaining a matching metric between the
3 two arbitrary images, the method comprising the steps of:

4 (a) obtaining filtered images using predetermined filters having different
5 orientation and scale coefficients with respect to the two arbitrary images; and

6 (b) obtaining a matching metric by calculating a minimum value of a sum of
7 absolute differences between variances and means of pixel values with respect to an
8 arbitrary filtered image and images filtered by filters having scale and orientation
9 coefficients different from those of filters used for filtering the arbitrary image.

ABSTRACT

A digital image texture analyzing method including the steps of (a) obtaining the mean (μ_0) and variance (σ_0) of the pixel values of an original image, and (b) obtaining $m \times n$ filtered images by filtering the original image using predetermined filters, each having a unique combination of one of m scales and one of n orientations, where m and n are predetermined positive integers, (c) calculating the means ($\mu_{11}, \mu_{21}, \dots, \mu_{mn}$) and variances ($\sigma_{11}, \sigma_{21}, \dots, \sigma_{mn}$) of the respective filtered images, and (d) obtaining a texture descriptor having the mean (μ_0) and variance (σ_0) of the pixel values of the original image obtained in the step (a), and the means ($\mu_{11}, \mu_{21}, \dots, \mu_{mn}$) and variances ($\sigma_{11}, \sigma_{21}, \dots, \sigma_{mn}$) of the respective filtered images obtained in the step (c), as texture features. The texture analyzing method allows image textures to be more accurately analyzed and compared. Also, according to the digital image texture analyzing method, when an image is only rotated, enlarged or reduced relative to another image, the similarity of the textures of two images are analyzed, that is, accurate analysis can be performed.

FIG. 1 (PRIOR ART)

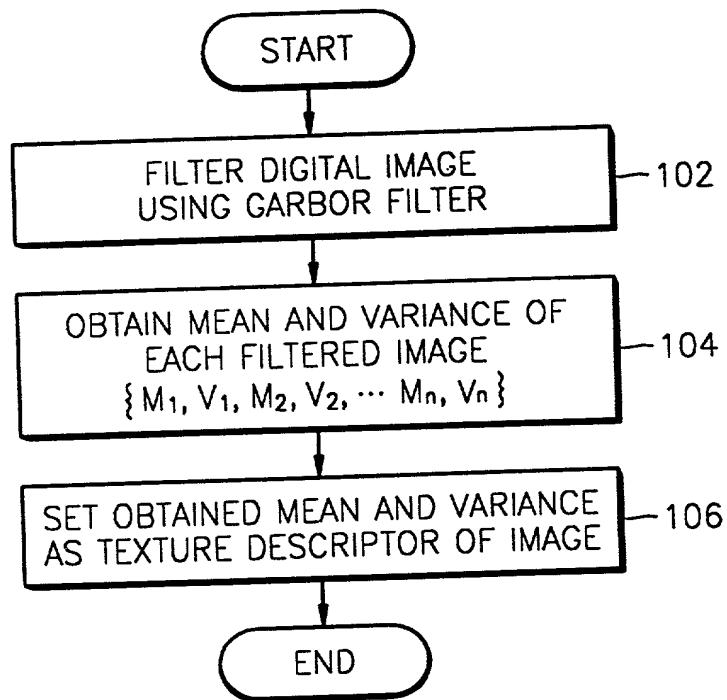


FIG. 2

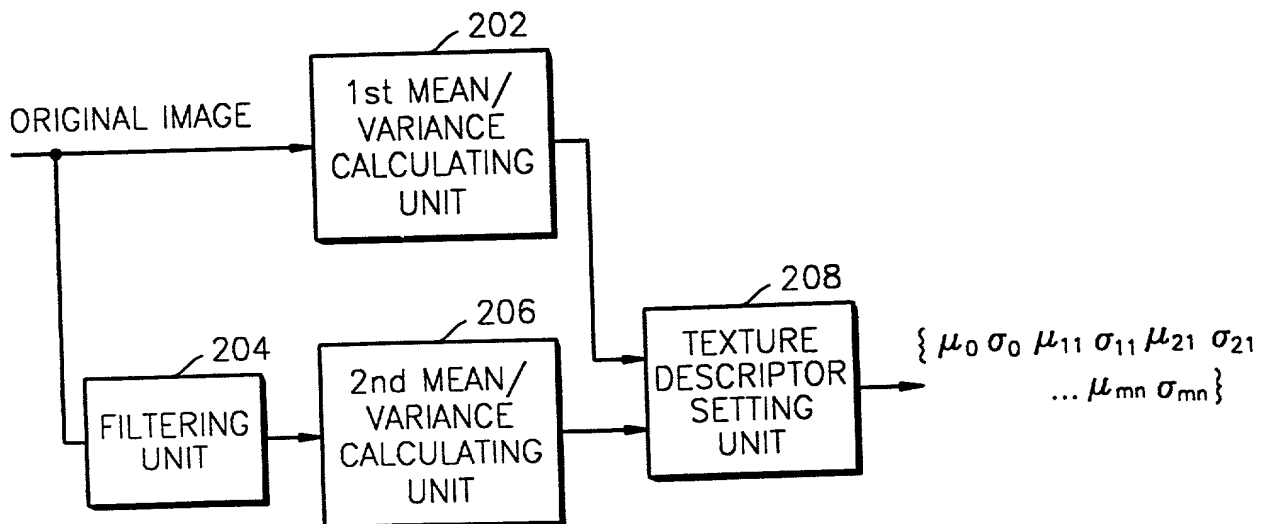


FIG. 3

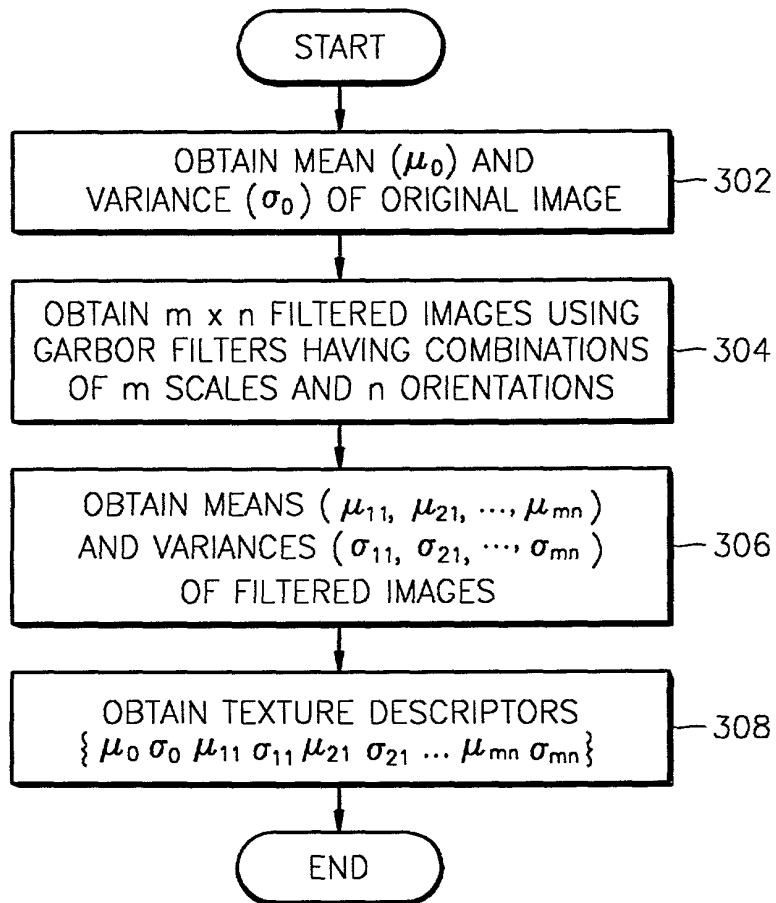


FIG. 4

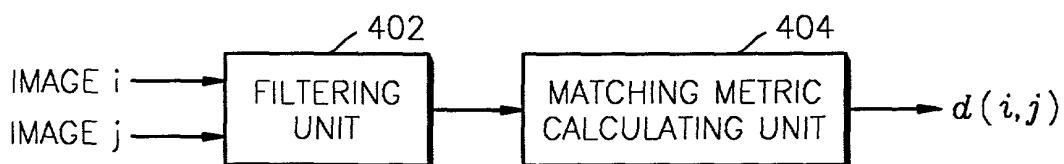


FIG. 5

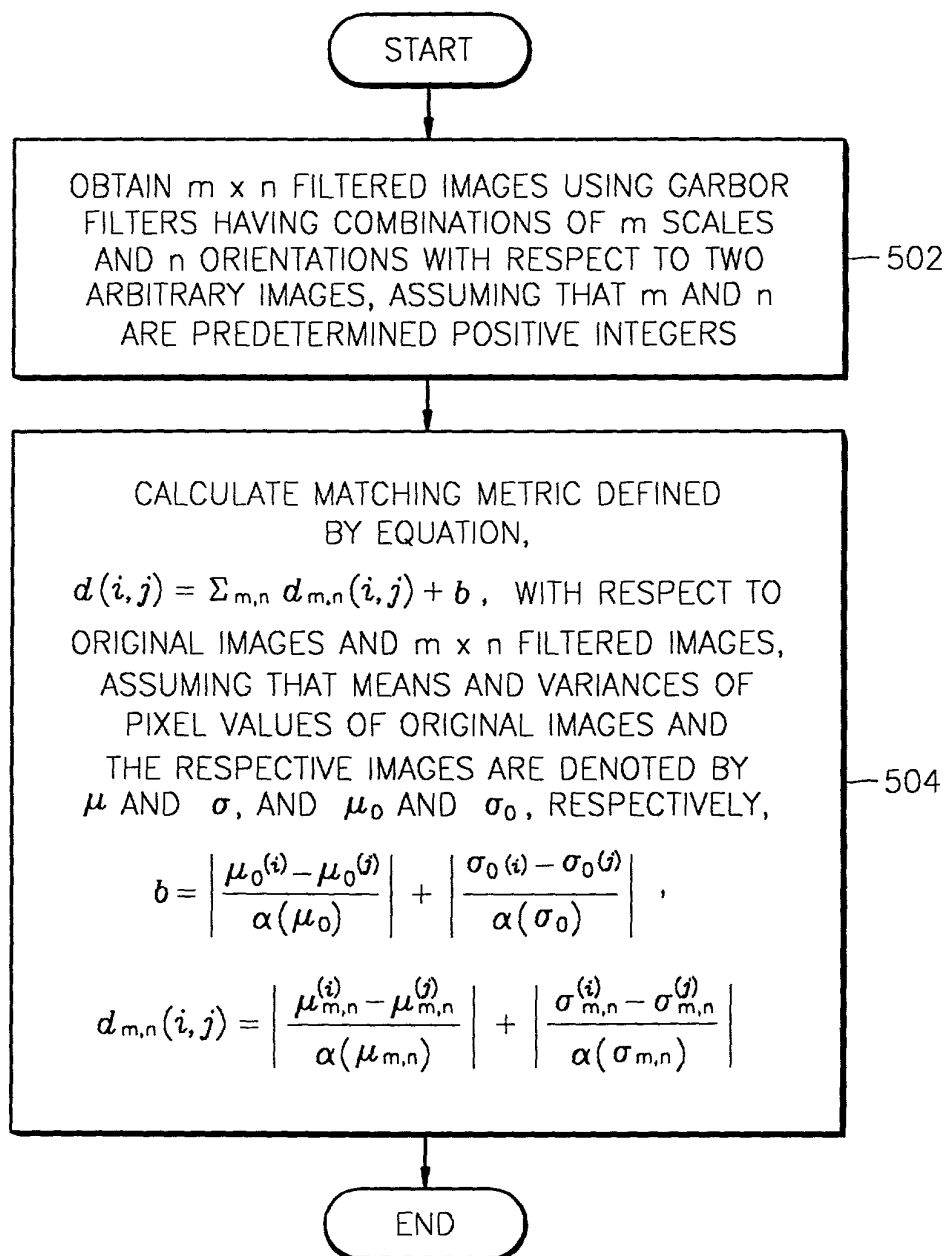


FIG. 6

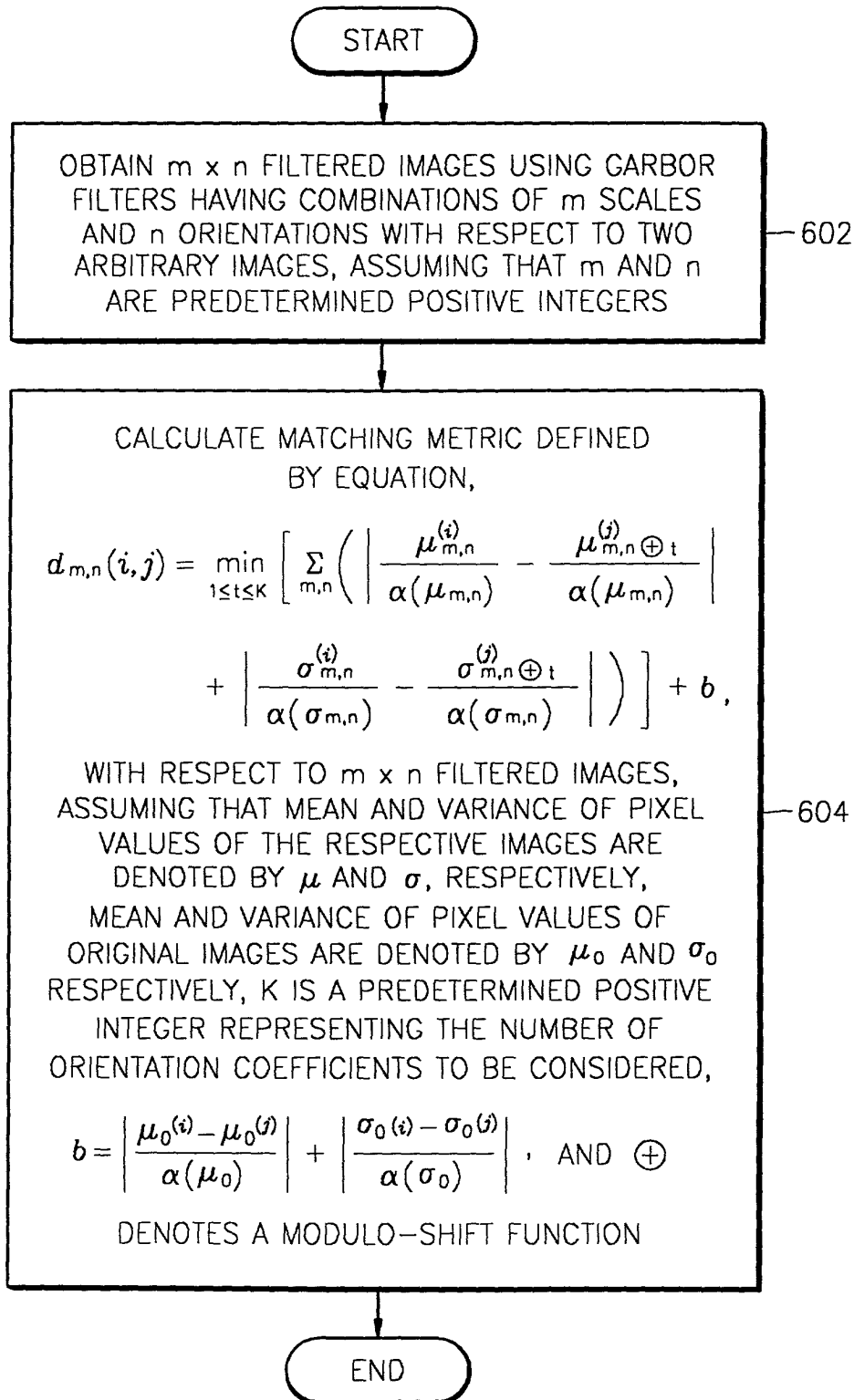


FIG. 7

